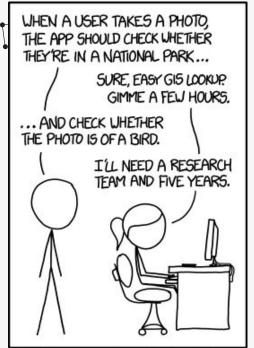
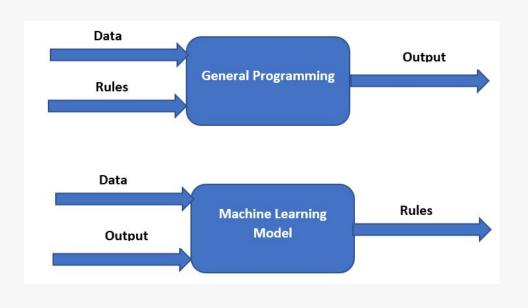
# Introduction to ML

Invento Research Inc.

# What is Machine Learning



IN CS, IT CAN BE HARD TO EXPLAIN THE DIFFERENCE BETWEEN THE EASY AND THE VIRTUALLY IMPOSSIBLE.



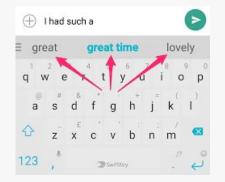


## ML examples











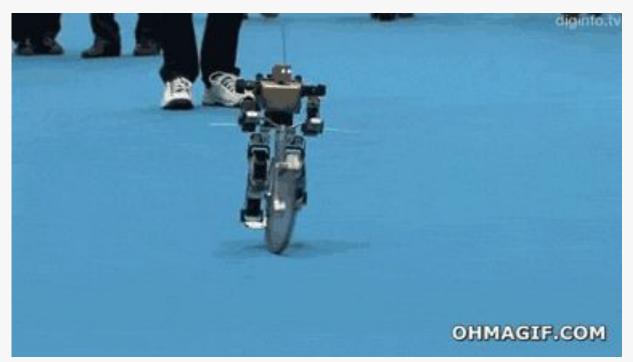






# Why Machine Learning?

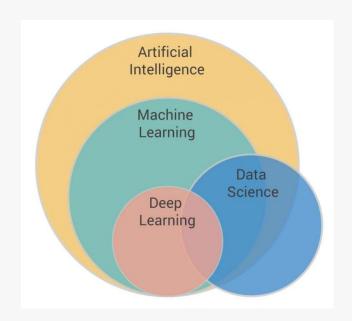


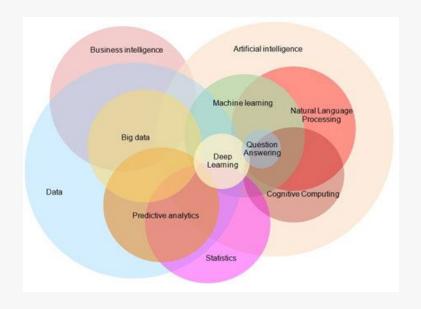




# Venn diagram of ML and deep learning

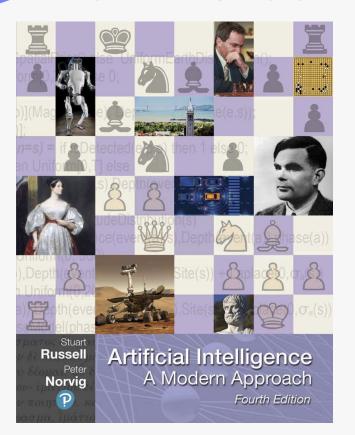








## Gurus of GOFAI





Stuart Russell and Peter Norvig



## The idea of AI is not new



#### 1956 Dartmouth Conference: The Founding Fathers of AI



John MacCarthy



**Marvin Minsky** 



Claude Shannon



**Ray Solomonoff** 



Alan Newell



**Herbert Simon** 



**Arthur Samuel** 



Oliver Selfridge



**Nathaniel Rochester** 

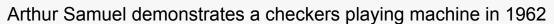


**Trenchard More** 



## ....neither is ML

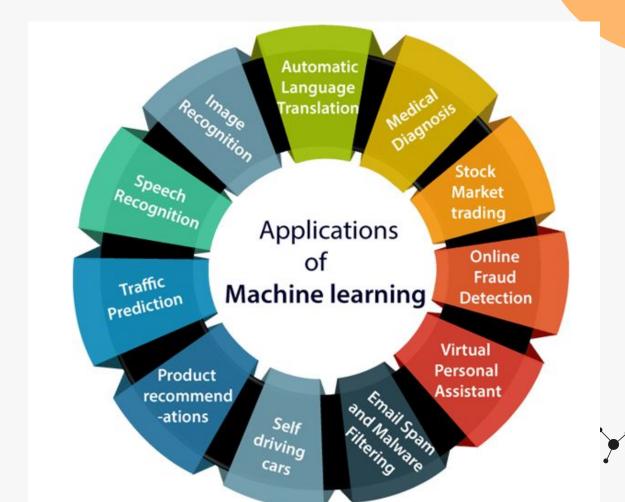




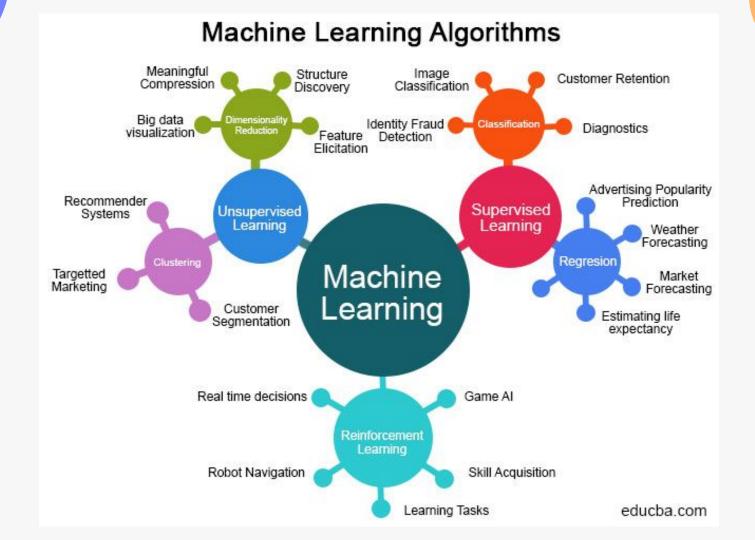


## Al timeline **ARTIFICAL INTELLIGENCE** MACHINE LEARNING stirs excitement. DEEP LEARNING Machine Learning Deep Learning begins to breakthroughs drive flourish. Al boom.

# Applied ML







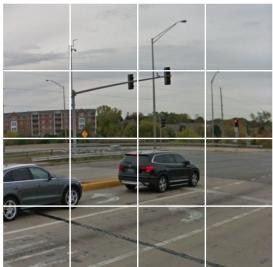


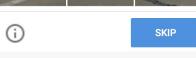
# ML - Supervised Learning

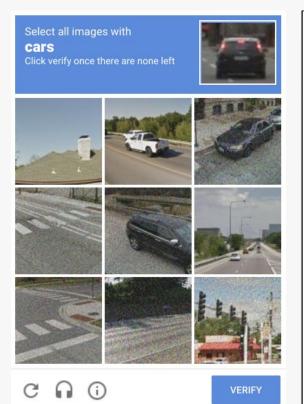
Select all squares with

traffic lights

If there are none, click skip





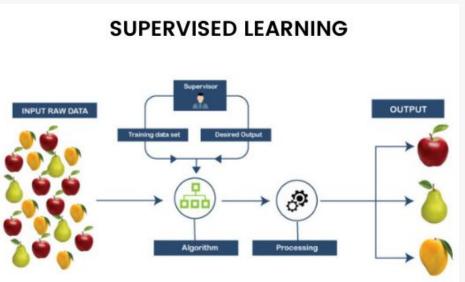


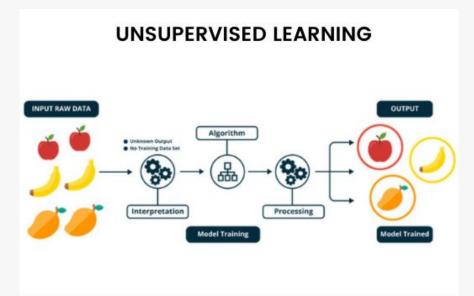


# **Unsupervised Learning**



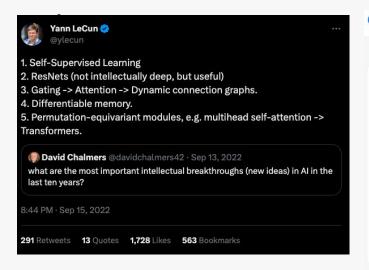
Introducing the concept of Similarity...!







# Self Supervised Learning



Meta Al

Research

# Self-supervised learning: The dark matter of intelligence

March 4, 2021

Implicit labels!

In recent years, the AI field has made tremendous progress in developing AI systems that can learn from massive amounts of carefully labeled data. This paradigm of supervised learning has a proven track record for training specialist models that perform extremely well on the task they were trained to do. Unfortunately, there's a limit to how far the field of AI can go with supervised learning alone.

## What is a model?







## What is a ML model?





## Parametric Models



Parameters are knobs!
In the linear model we have 2 parameters - slope and intercept

Parametric models can have many parameters - sometimes in the billions! (Like ChatGPT)



Non-Parametric models like trees will be covered in later slides



# Predicting house prices







## Data and Features

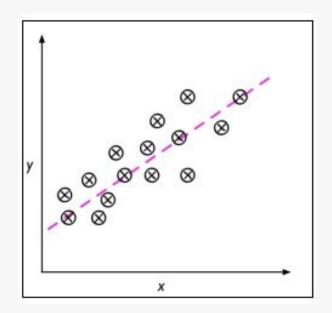


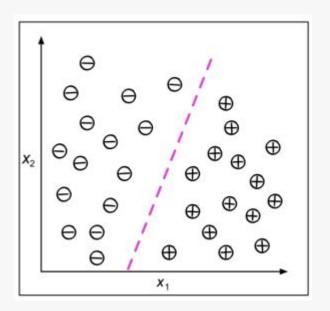
	0	1	2	3	4
id	7129300520	6414100192	5631500400	2487200875	1954400510
date	10/13/2014	12/9/2014	2/25/2015	12/9/2014	2/18/2015
price	221900	538000	180000	604000	510000
bedrooms	3	3	2	4	3
bathrooms	1	2.25	1	3	2
sqft_living	1180	2570	770	1960	1680
sqft_lot	5650	7242	10000	5000	8080
floors	1	2	1	1	1
waterfront	0	0	0	0	0
view	0	0	0	0	0
condition	3	3	3	5	3
grade	7	7	6	7	8
sqft_above	1180	2170	770	1050	1680
q <mark>ft_base</mark> ment	0	400	0	910	0
yr_built	1955	1951	1933	1965	1987
yr_renovated	0	1991	0	0	0
zipcode	98178	98125	98028	98136	98074
lat	47.5112	47.721	47.7379	47.5208	47.6168
long	-122.257	-122.319	-122.233	-122.393	-122.045
sqft_living15	1340	1690	2720	1360	1800
sqft_lot15	5650	7639	8062	5000	7503



# Regression vs Classification









# Setting up the work environment



- Introduction to Jupyter
- Google Colab [https://colab.research.google.com/]



# Drawing a line



#### THE EQUATION FOR THE LINE:

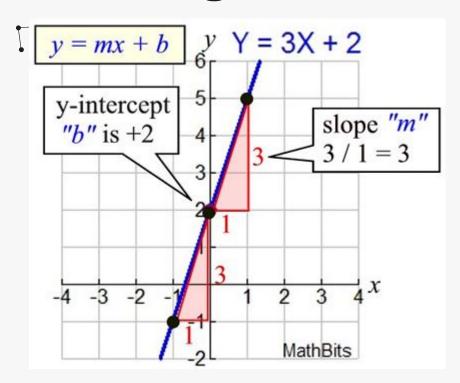
$$y = mx + b$$

$$y = INTERCEPT$$

$$m=1 \\ b=5 \qquad \Rightarrow \qquad y=1 \times +5 = x+5$$



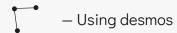
# Drawing a line in 2D



Slope = Rise over run

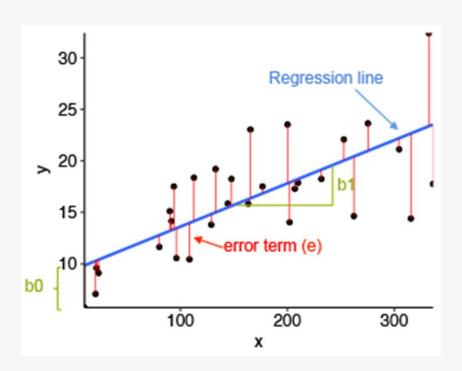


# Let's look at a live example





# How good is our prediction?





#### **Bad Predictions**

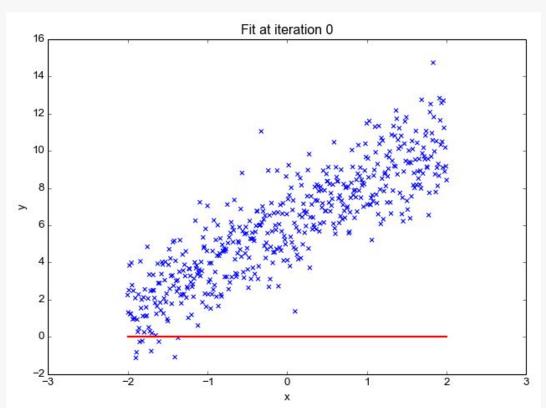


MSE = 
$$\frac{1}{n} \sum_{i=1}^{n} (y_i - \tilde{y}_i)^2$$

Mean squared error cost function



# Training iteratively



MSE = 
$$\frac{1}{n} \sum_{i=1}^{n} (y_i - \tilde{y}_i)^2$$

Training a model = Minimizing the loss (or cost)

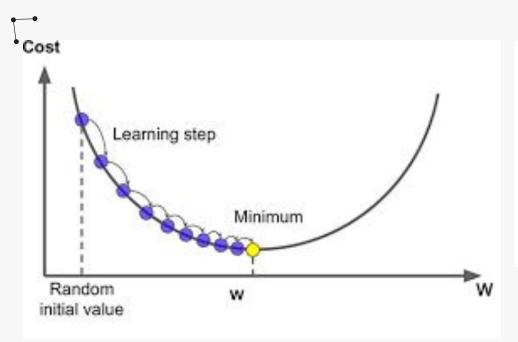


## xkcd - ML

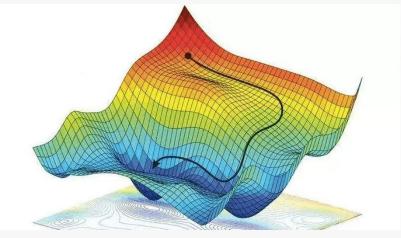
# THIS IS YOUR MACHINE LEARNING SYSTEM? YUP! YOU POUR THE DATA INTO THIS BIG PILE OF LINEAR ALGEBRA, THEN COLLECT THE ANSWERS ON THE OTHER SIDE. WHAT IF THE ANSWERS ARE WRONG? JUST STIR THE PILE UNTIL THEY START LOOKING RIGHT.



# Plotting the loss function



One Parameter MSE loss function



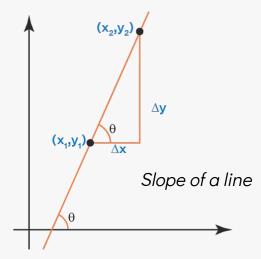
Multi Parameter loss function



# Gradient Descent - the magic behind Al



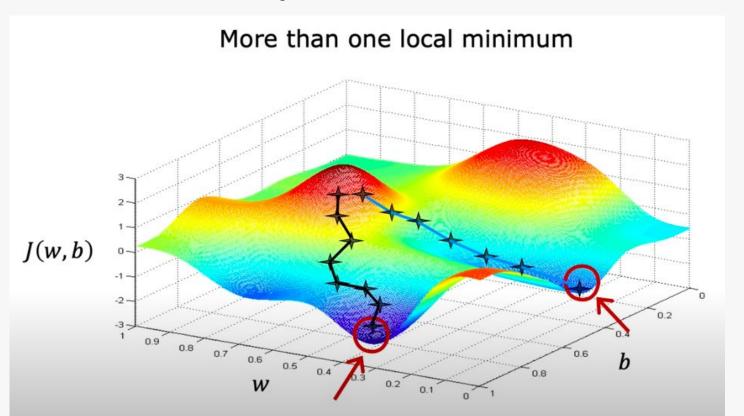
Using differential calculus to make tiny adjustments to the knobs automatically!





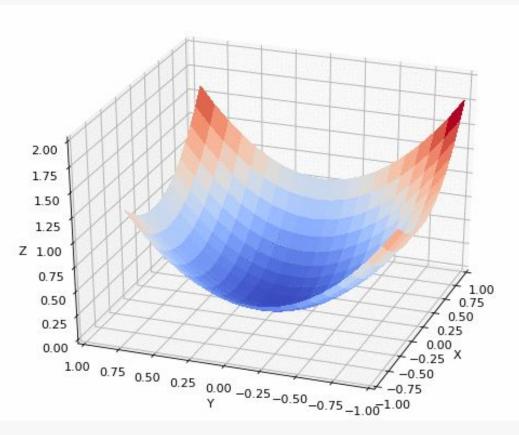
# The loss landscape

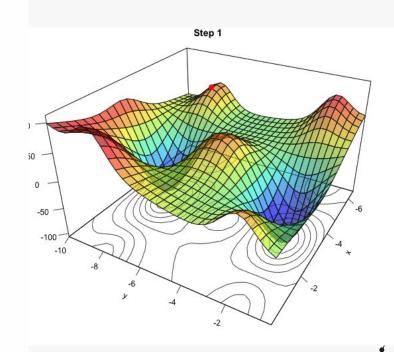






## Iterative training - step by step adjustments





## Gradient Descent - derivatives



#### Linear regression model

#### Cost function

$$f_{w,b}(x) = wx + b$$
  $J(w,b) = \frac{1}{2m} \sum_{i=1}^{m} (f_{w,b}(x^{(i)}) - y^{(i)})^2$ 

#### Gradient descent algorithm

repeat until convergence {

$$w = w - \alpha \frac{\partial}{\partial w} J(w, b) \qquad \frac{1}{m} \sum_{i=1}^{m} (f_{w,b}(x^{(i)}) - y^{(i)}) x^{(i)}$$

$$b = b - \alpha \frac{\partial}{\partial b} J(w, b) \qquad \frac{1}{m} \sum_{i=1}^{m} (f_{w,b}(x^{(i)}) - y^{(i)})$$



# Introduction to Logistic regression



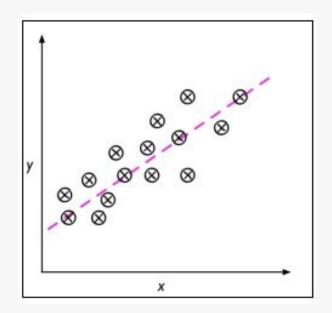
Don't let the name confuse you. For historical reasons it is called 'regression'.

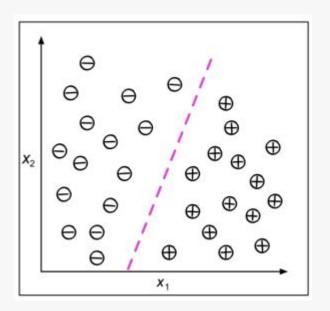
Logistic regression is a *CLASSIFICATION* algorithm!



# Regression vs Classification



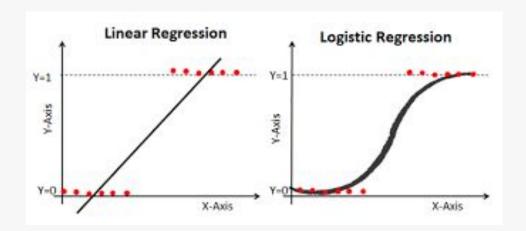


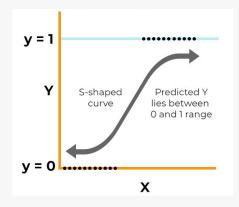




# From straight lines to curves



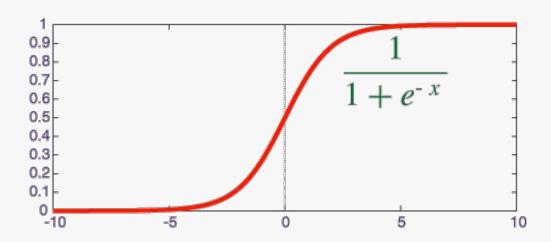






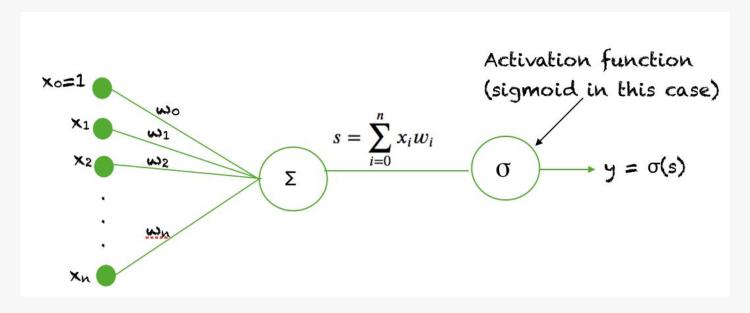
# Sigmoid Function







# 





# Cost function for logistic regression

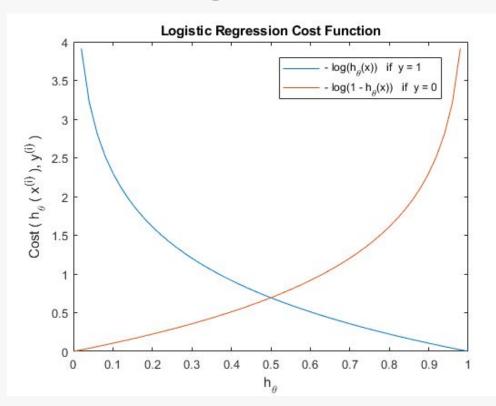


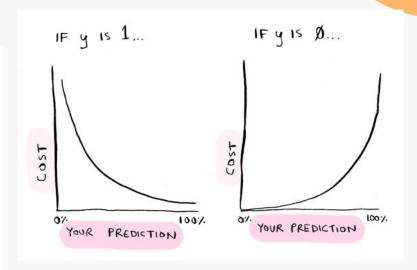
$$J(\theta) = \frac{1}{m} \sum_{i=1}^{m} Cost(h_{\theta}(x^{(i)}), y^{(i)})$$

$$J(\theta) = \frac{1}{m} \left[ \sum_{i=1}^{m} -y^{(i)} log(h_{\theta}(x^{(i)})) + (1 - y^{(i)}) log(1 - h_{\theta}(x^{(i)})) \right]$$

m = number of samples

# Why logarithm?



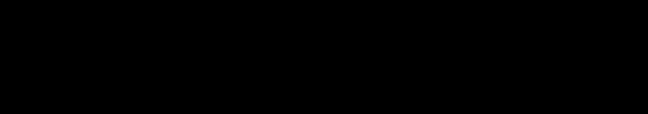


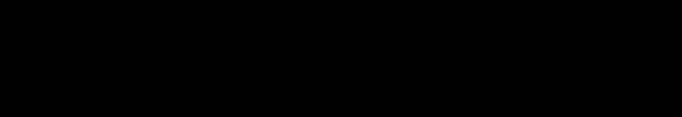
$$J(\theta) = \frac{1}{m} \sum_{i=1}^{m} Cost(h_{\theta}(x^{(i)}) - y^{(i)})$$

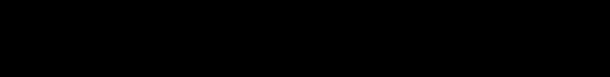
$$\operatorname{Cost}(h_{\theta}(x), y) = \begin{cases} -\log(h_{\theta}(x)) & \text{if y=1} \\ -\log(1 - h_{\theta}(x)) & \text{if y=0} \end{cases}$$

# Log loss calculation





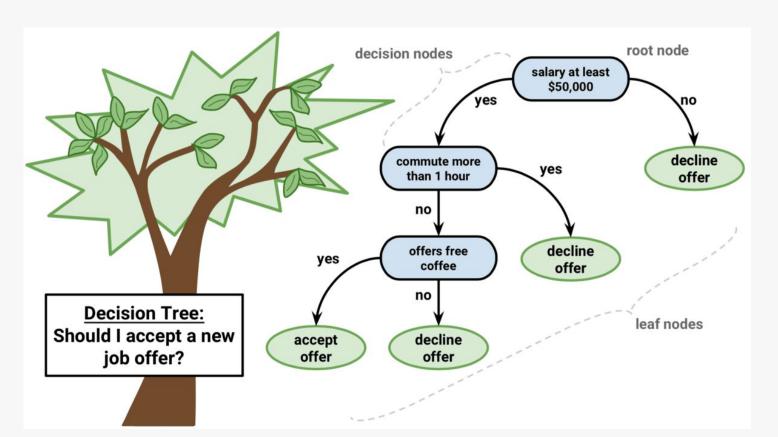






## Non-Parametric models

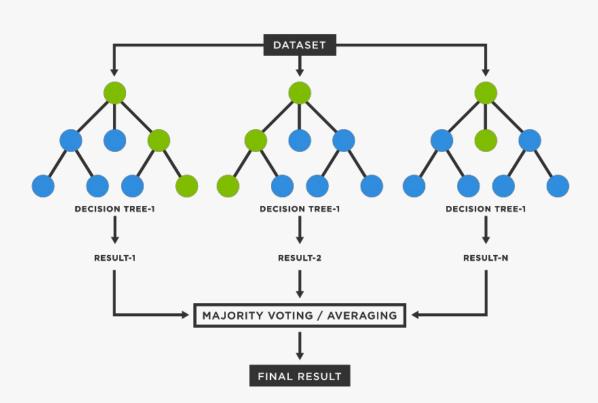






## Ensemble models









## Module 2

